

F-15 837 IFCS Intelligent Flight Control System Project

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NASA, Dryden Flight Research Center



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Project Participants



- **Nasa Dryden Flight Research Center**
 - Responsible test organization for the flight experiment
 - Flight, range and ground safety
 - Mission success
- **Nasa Ames Research Center**
 - Development of the concepts
- **Boeing STL Phantom Works**
 - Primary flight control system software (Conventional mode)
 - Research flight control system software (Enhanced mode)
- **Institute for Scientific Research**
 - Neural Network adaptive software
- **Academia**
 - West Virginia University
 - Georgia Tech
 - Texas A&M



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F-15 IFCS Project Goals

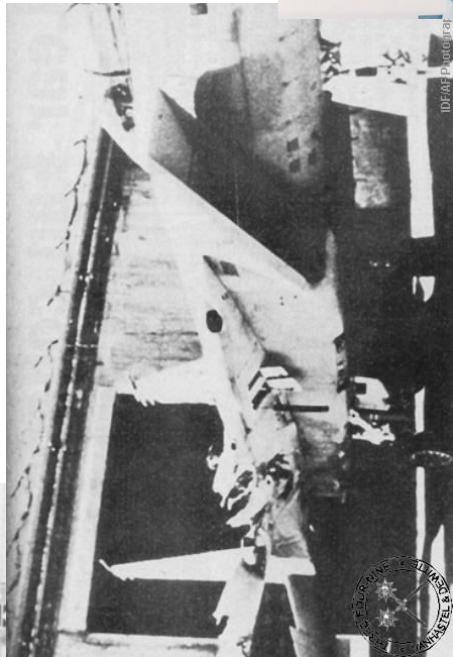
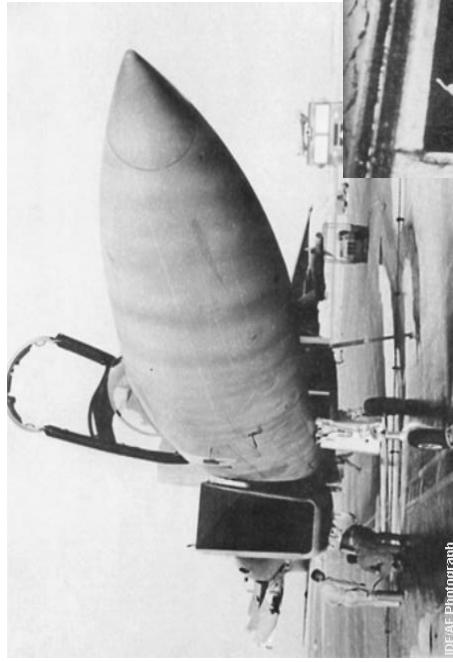
- Demonstrate Revolutionary Control Approaches that can Efficiently Optimize Aircraft Performance in both Normal and Failure Conditions
- Advance Neural Network-Based Flight Control Technology for New Aerospace Systems Designs



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Motivation

These are survivable accidents



**IFCS has potential to
reduce the amount of
skill and luck required
for survival**

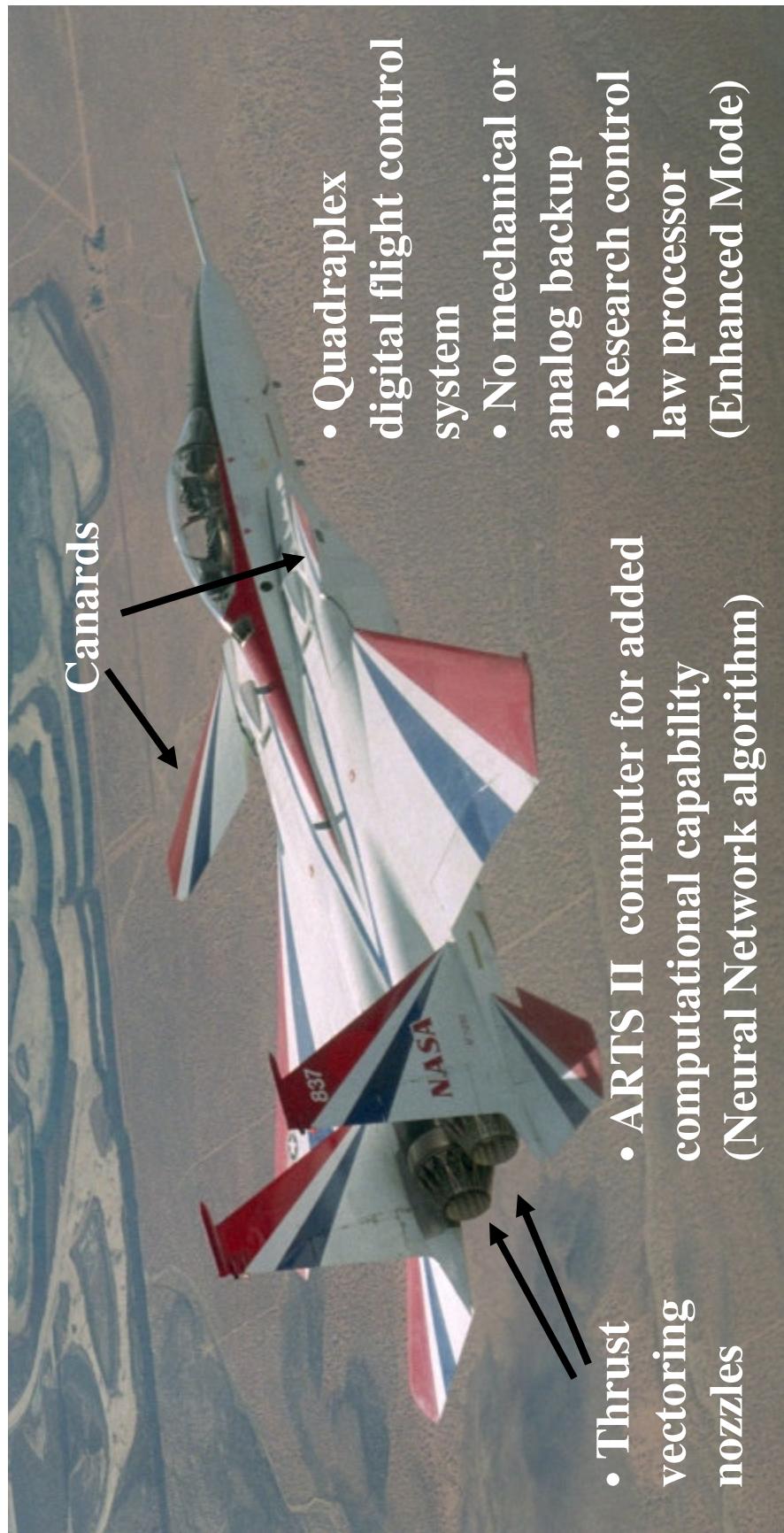


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NASA NF-15B Tail Number 837

Extensively modified F-15 airframe

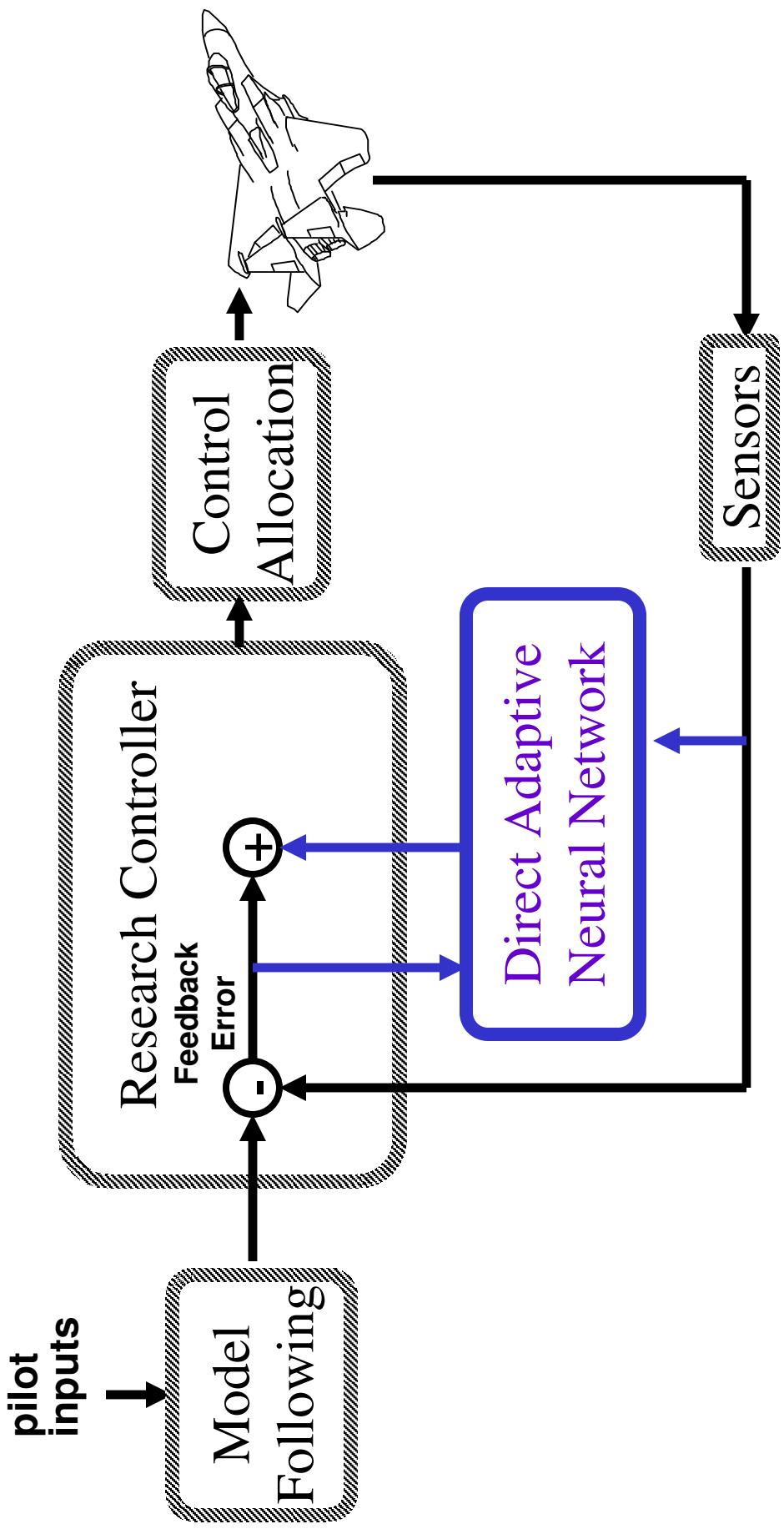


- Quadraplex digital flight control system
- No mechanical or analog backup
- Research control law processor (Enhanced Mode)
- ARTS II computer for added computational capability (Neural Network algorithm)
- Thrust vectoring nozzles



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Gen II Direct Adaptive Control Architecture (Adaptive)



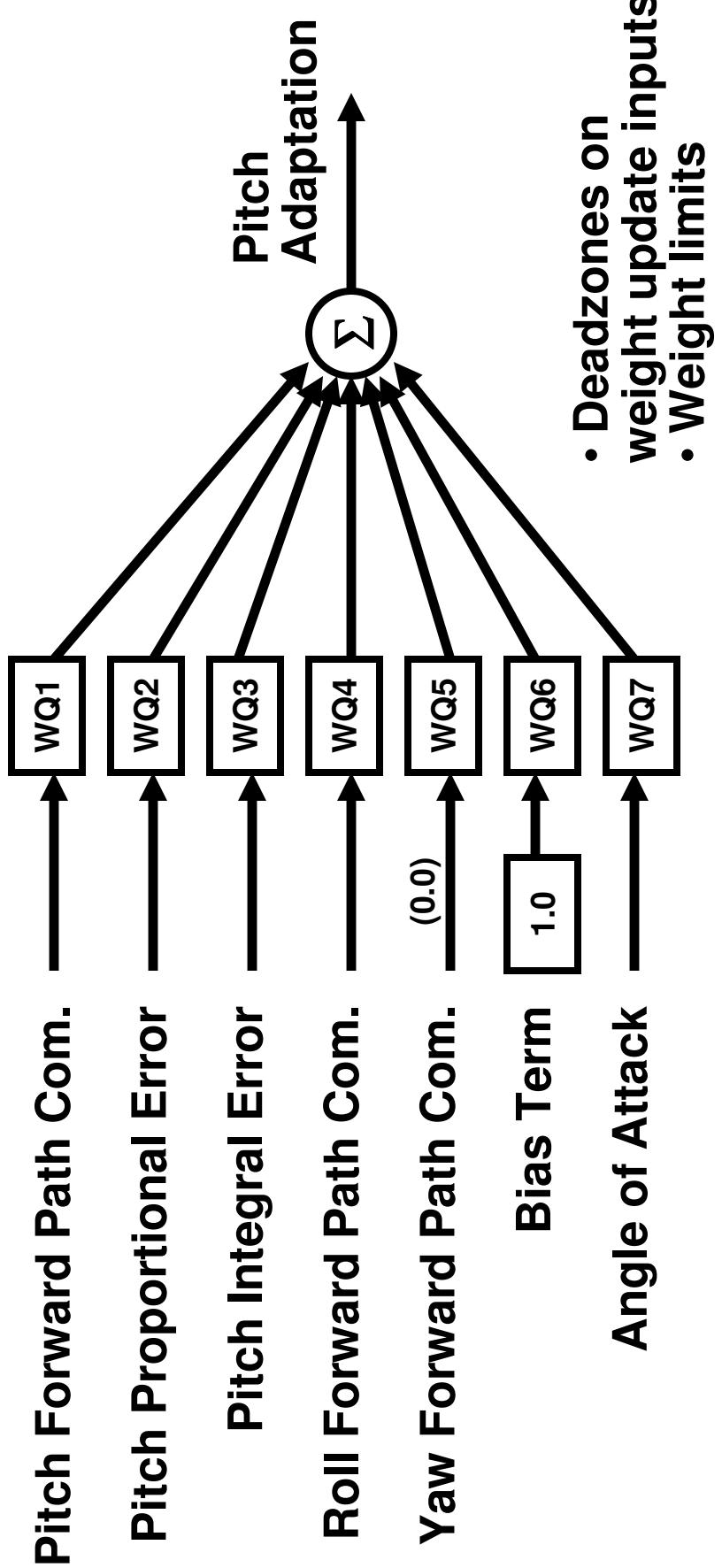
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Simplified Sigma-Pi Neural Network

Pitch Axis



Weight Update Law: $\dot{W} = -G(U_{err} B_a + L U_{err} W) dt$



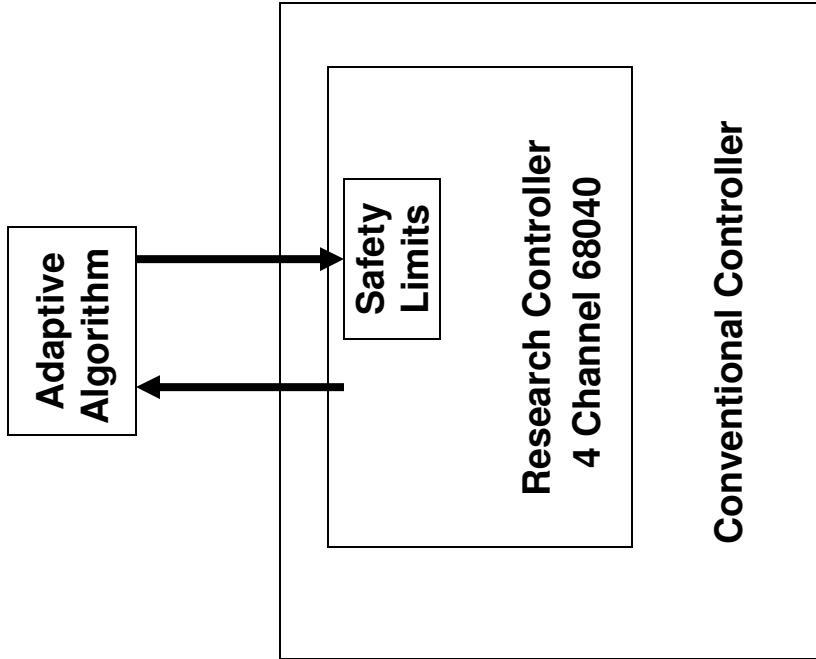
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Limited Authority System

- Adaptation algorithm implemented in separate processor
 - Class B software
 - Autocoded directly from Simulink block diagram
 - Many configurable settings
 - Learning rates
 - Weight limits
 - Thresholds, etc.
- Control laws programmed in Class A, quad-redundant system
- Protection provided by floating limiter on adaptation signals

Single Channel 400 MHz



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Flight Experiment

- **Assess handling qualities of Gen II controller without adaptation**
- **Activate adaptation and assess changes in handling qualities**
- **Introduce simulated failures**
 - Control surface locked (“B matrix failure”)
 - Angle of attack to canard feedback gain change (“A matrix failure”)
- **Re-assess handling qualities with simulated failures and adaptation.**
- **Report on “Real World” experience with a neural network based flight control system**



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Adaptation Goals

- Ability to suppress initial transient due to failure
 - Trade-off between high learning rate and stability of system
- Ability to re-establish model following performance
- Ability to suppress cross coupling between axes
 - No existing criteria



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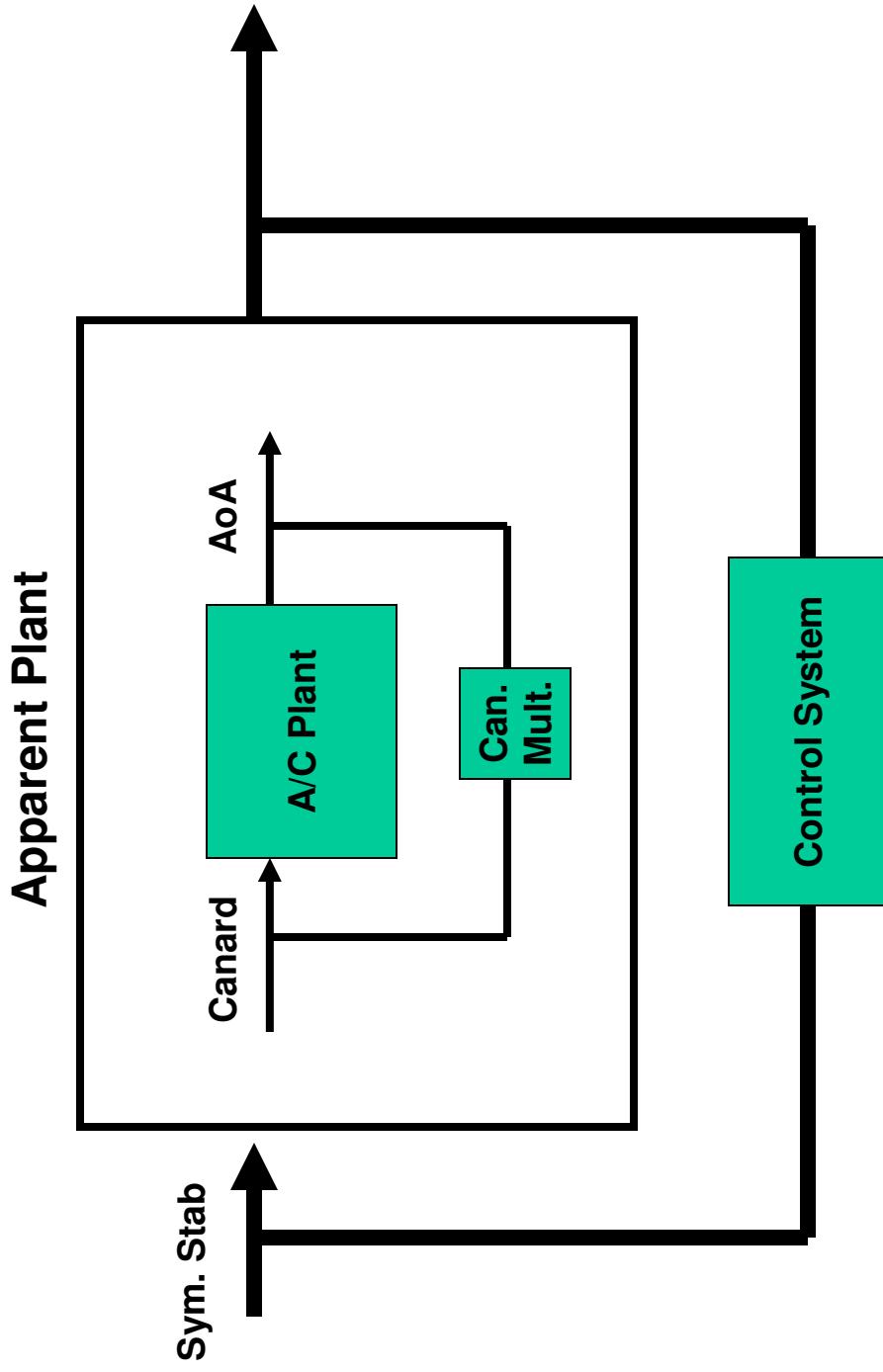


Simulated Destabilization A-Matrix Failure



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Effect of Canard Multiplier



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Canard Multiplier Effect Closed Loop Freq. Resp.

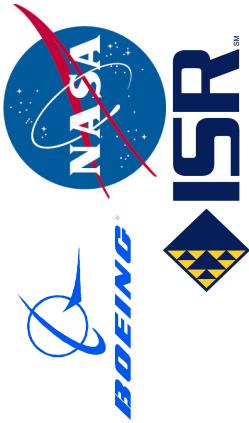
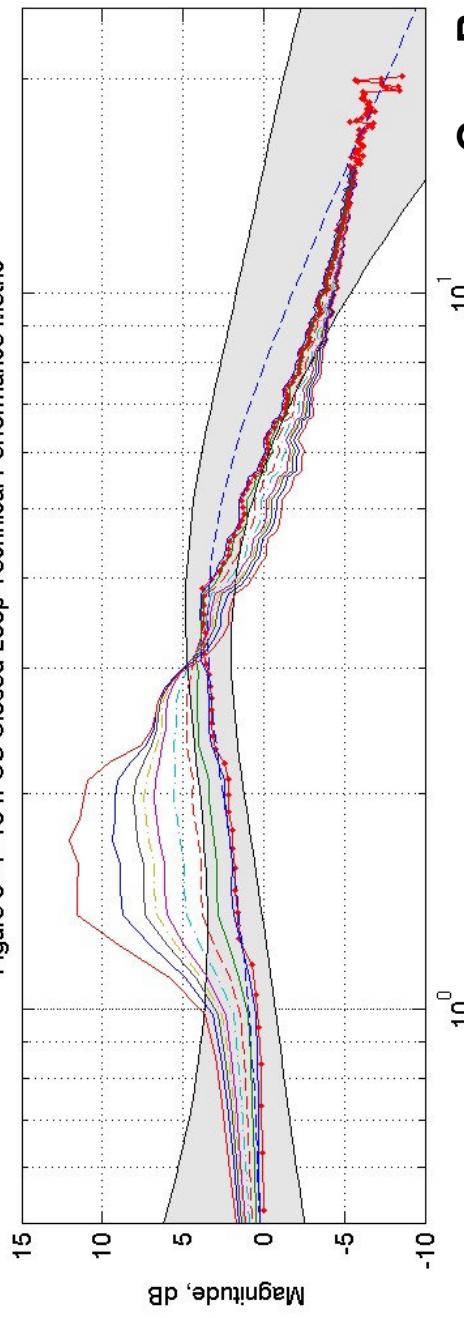
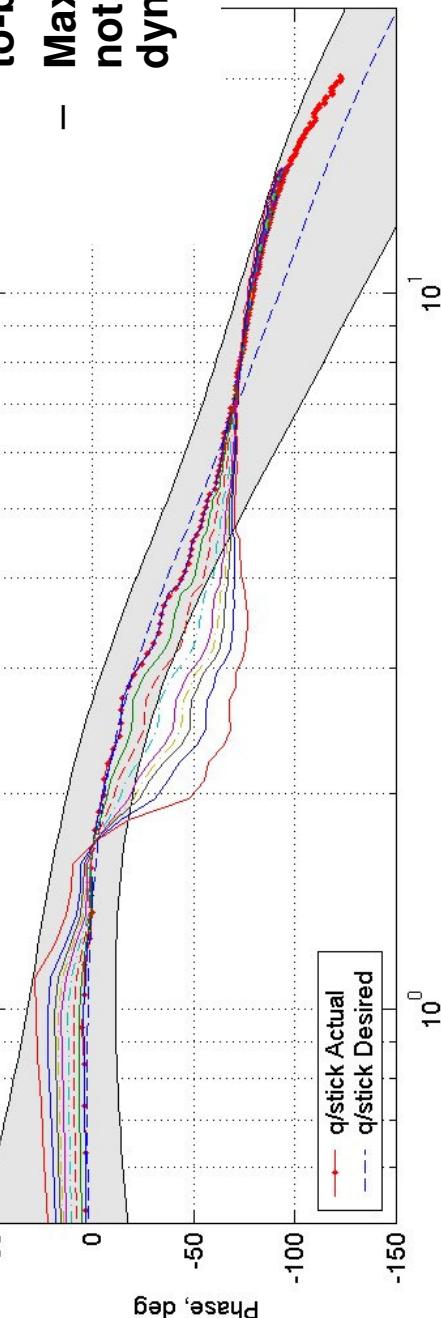


Figure 3 - F-15 IFCS Closed Loop Technical Performance Metric



• Grey Region:

- Based on model-to-be-followed
- Maximum noticeable dynamics (LOES)



Frequency, rad/sec

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Canard Multiplier Effect Closed Loop with Adaptation

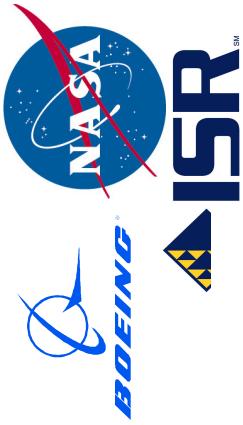
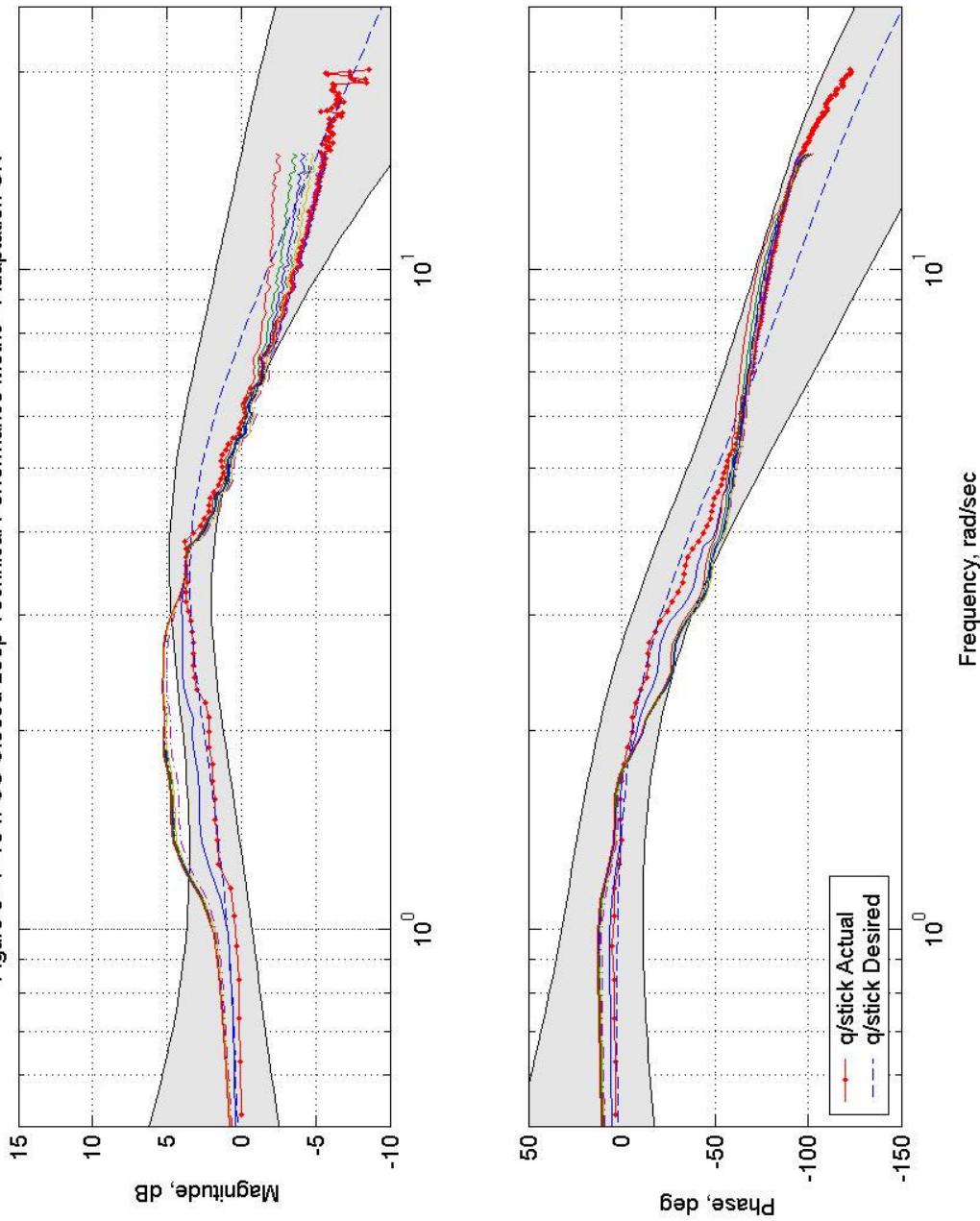
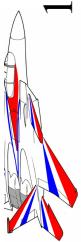


Figure 5 - F-15 IFCS Closed Loop Technical Performance Metric - Adaptation ON



Frequency, rad/sec

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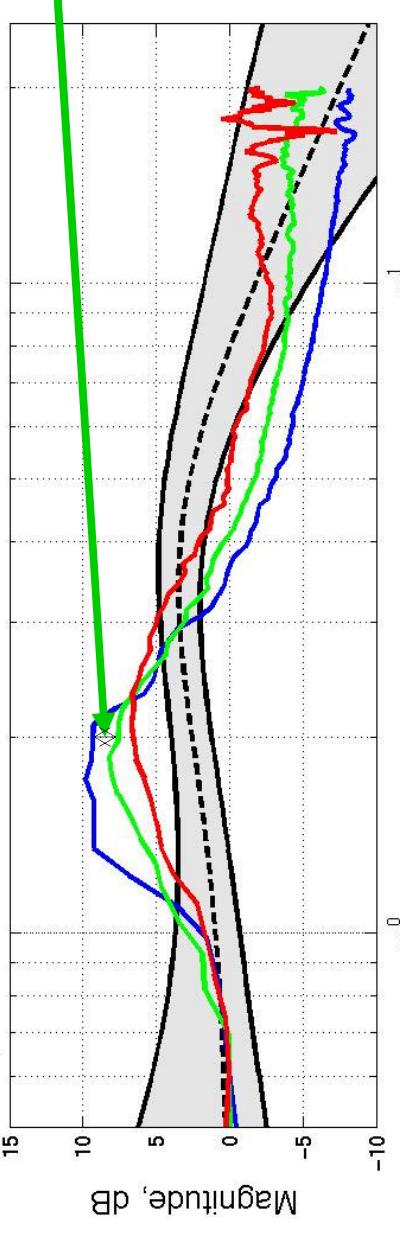




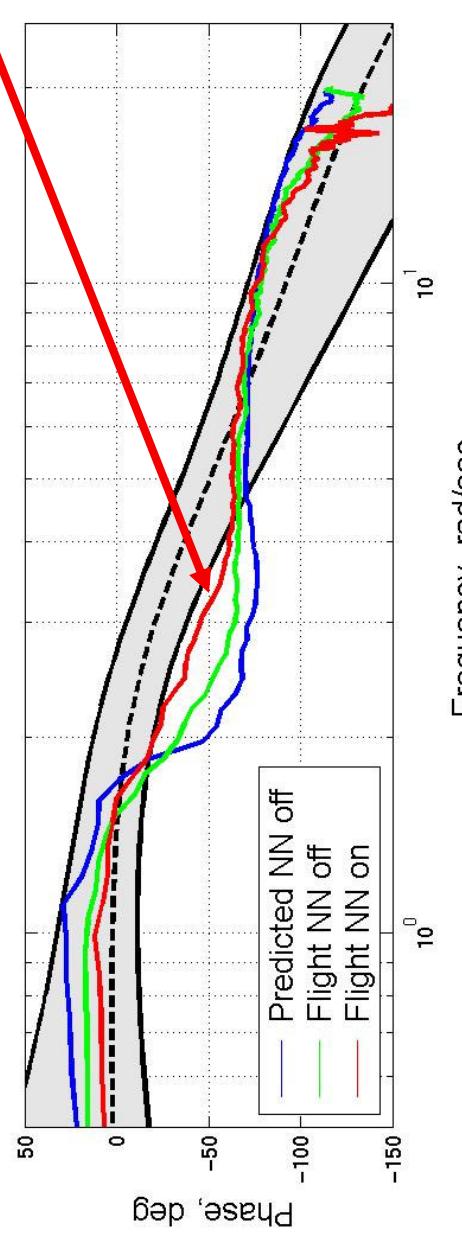
Simulated Destabilization Failure (Angle of Attack Feedback Change)



Closed Loop Pitch Axis Technical Performance Metric $M=0.75$ $H=20K$ $CM=-0.5$



- Effect of simulated failure less than predicted



- Adaptation

Improved response

- Software change in work to increase failure size



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Simulated Stabilator Failure



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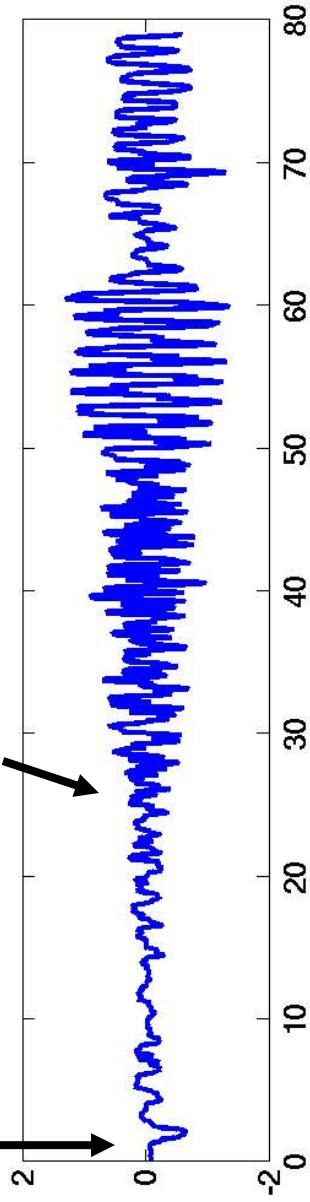


Adaptation Time History

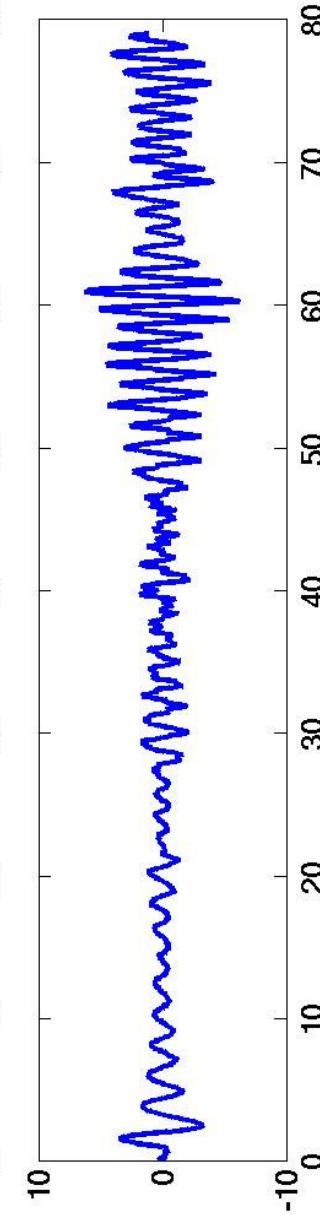
Failure Insertion

Tracking Starts

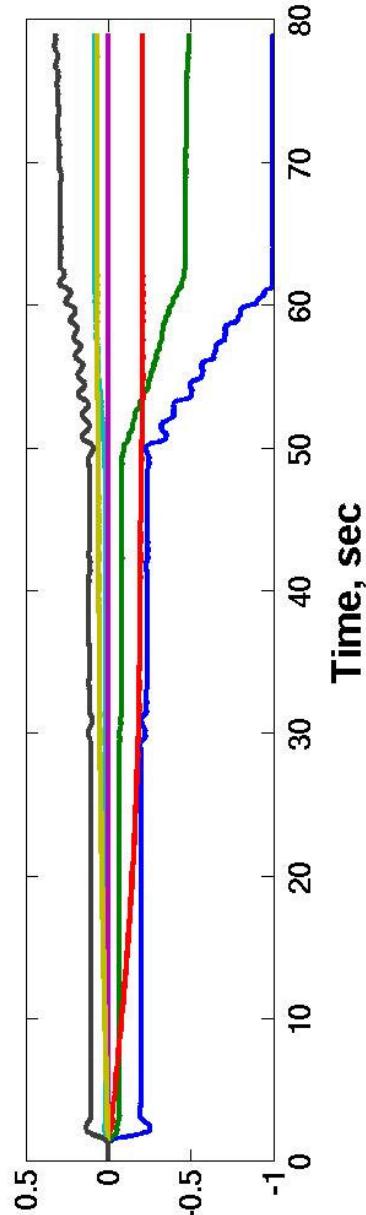
Long. Stick,
in.



Pitch Rate,
deg/sec



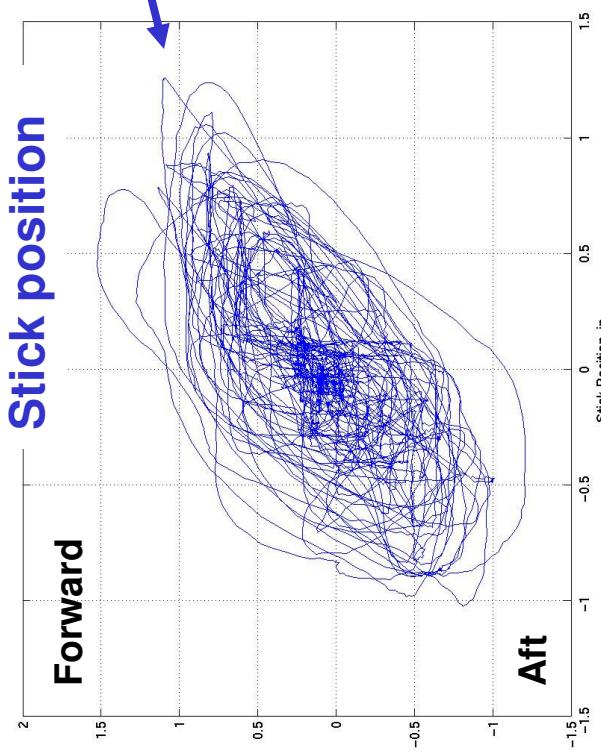
Normalized
pitch
weights



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Simulated Frozen Stabilator

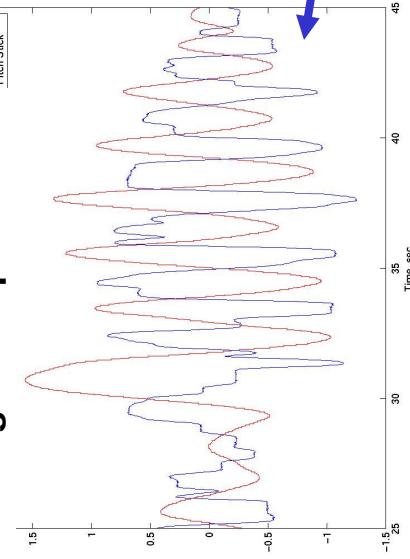


- Pilot unconsciously compensates for asymmetry

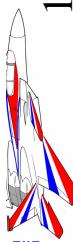
- Correlated pilot input presents greater challenge for adaptive system

- Adaptive system reduced the amount of cross coupling

180 deg out of phase



- Adaptive system also introduced tendency for pilot induced oscillations (PIO)



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Deadzone Effect

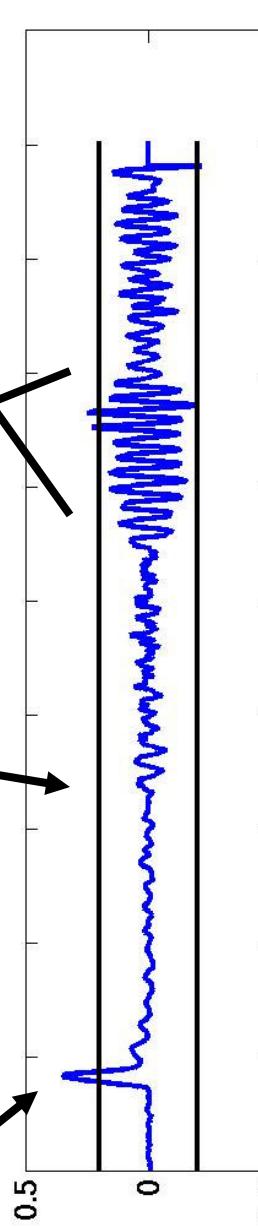


Failure Insertion

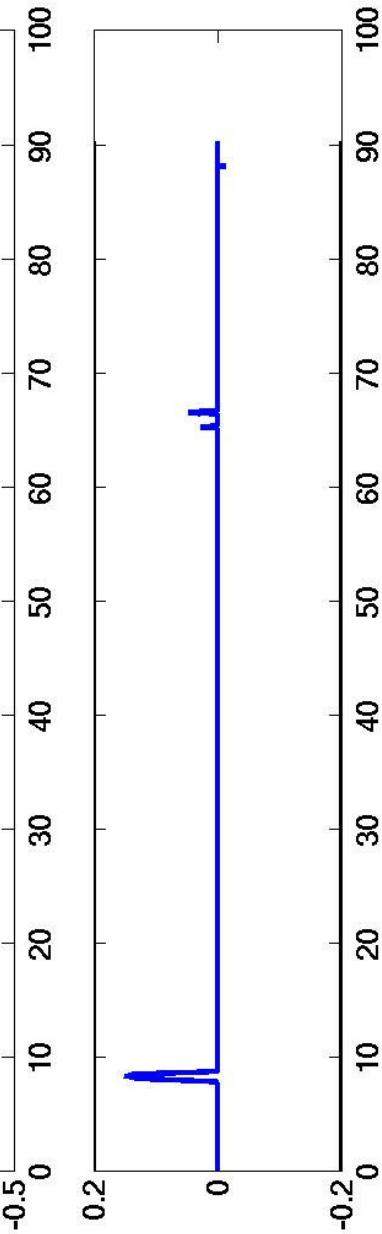
Tracking Starts

PIO

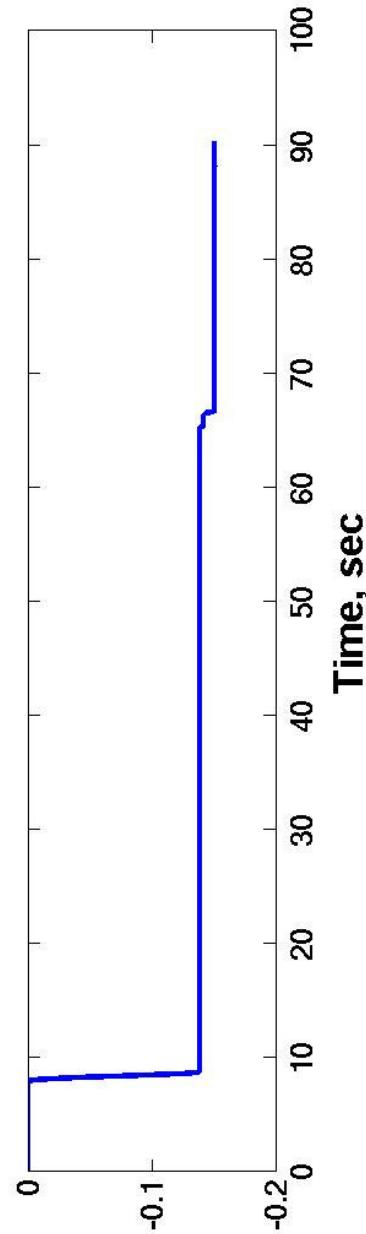
Input to
WP1 learning



After
deadzone



WP1



Time, sec



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Direct Adaptive Experience and Lessons Learned



- Initial simulation model had high bandwidth
 - Majority of system performance achieved by the dynamic inversion controller
 - Direct adaptive NN played minor role
- Dynamic Inversion gains reduced to meet ASE attenuation requirements
 - Much harder to achieve desired performance
 - NN contribution increased
- Initial performance objective emphasized transient reduction and achieving model following after failure
 - Piloted simulation results showed that reducing cross coupling was more important objective
- Explicit cross terms in NN required for failure cases
 - Relying on disturbance rejection alone doesn't work (also finding of Gen 1)



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Direct Adaptive Experience and Lessons Learned



- **Liapunov proof of bounded stability**
 - Necessary but not sufficient proof of stability
 - Many cases of limit cycle behavior observed
 - Other analytic methods required for ensuring global stability
- **Dynamic Inversion controller contributes significantly to cross coupled response in presence of surface failure (locked)**
 - Redesigned yaw loop using classical techniques
- **NN's require careful selection of inputs**
 - Presence of transient errors "normal" for abrupt inputs in non-adaptive systems
 - Existence of transient errors tend to drive NN's to "high gain" trying to achieve impossible
- **Significant amount of "tuning" required for to achieve robust full envelope performance**
 - Contradicts claim of robustness to unforeseen failures
 - Piloted nonlinear simulation required



Conclusions

- Adaptive system generally behaved as predicted
 - Weights adjusted in correct direction
 - Real world turbulence and measurement noise did not adversely affect learning
 - Only safety disengagements observed were due to very aggressive pilot inputs
- Simulated destabilization less than predicted
 - Flight vehicle more stable than aero model predicts
 - Software change in work to increase destabilizing gain
- No metrics currently exist for damaged vehicles
- Gained valuable real world experience that has already pushed technology to more acceptable level



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Potential Future Work



- How to sense and incorporate structural limitations into the adaptive algorithm
- Develop better metrics – What is most important to ensure that a damaged vehicle can be safely landed?
- Further investigation of asymmetric failure
 - Determine source of PIO and develop means to suppress it
 - Does stab failure require more complicated (nonlinear) neural network or direct adaptation of control surface mixer?
- Investigate adaptive notch filters to avoid adverse aero-servo-elastic (ASE) interactions
- Maintain long-term effort to advance adaptive control technology



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